



THE UNIVERSITY OF TEXAS AT AUSTIN
RADIONAVIGATION LABORATORY

UT  **SAVES**



Collaborative All-Weather Sensing for Automated Vehicles

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UT San Antonio | April 20, 2018



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RADIONAVIGATION LABORATORY

Ph.D. Track Students

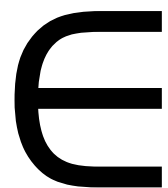


Undergraduate Research Assistants





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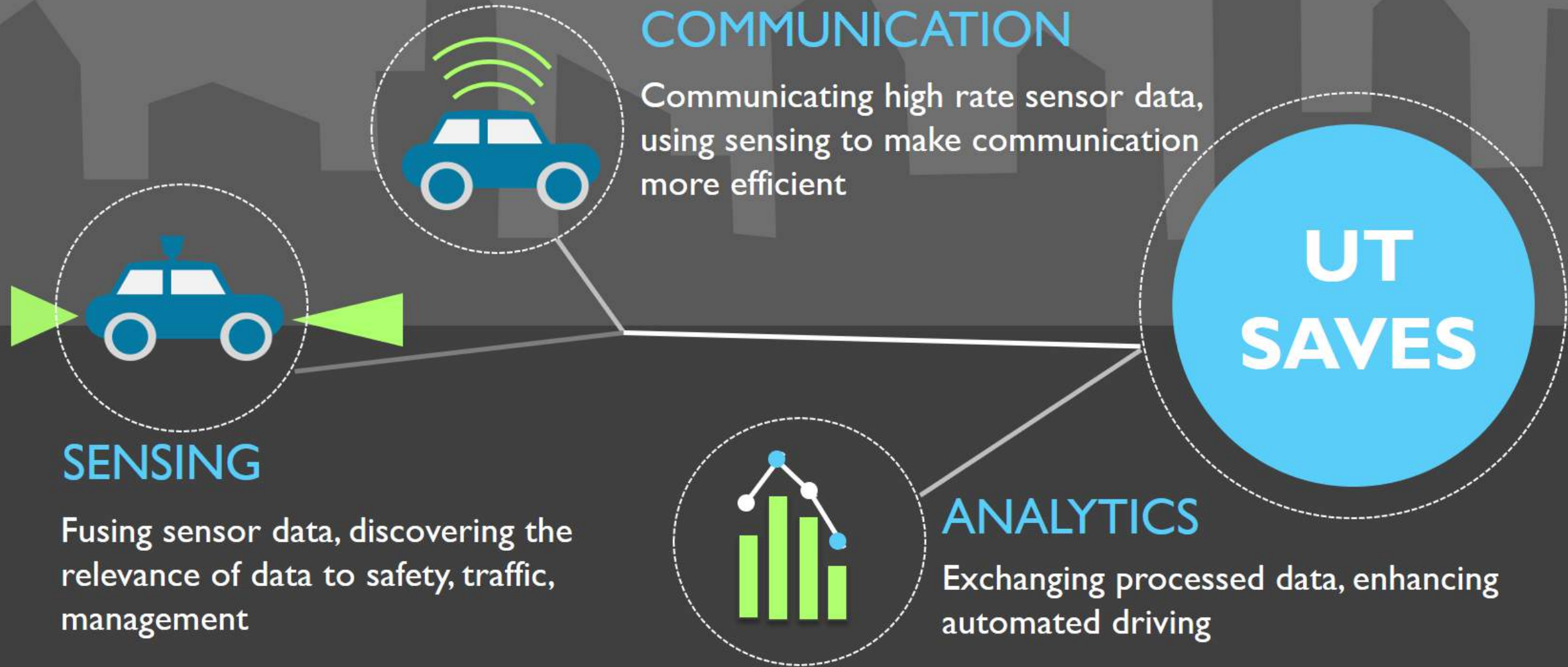


WELCOME TO UT SAVES

SAVES (Situation-Aware Vehicular Engineering Systems) is a research center within UT's Wireless Networking and Communications Group (WNCG) that addresses the challenges of wireless, networking, and sensing in vehicular systems.

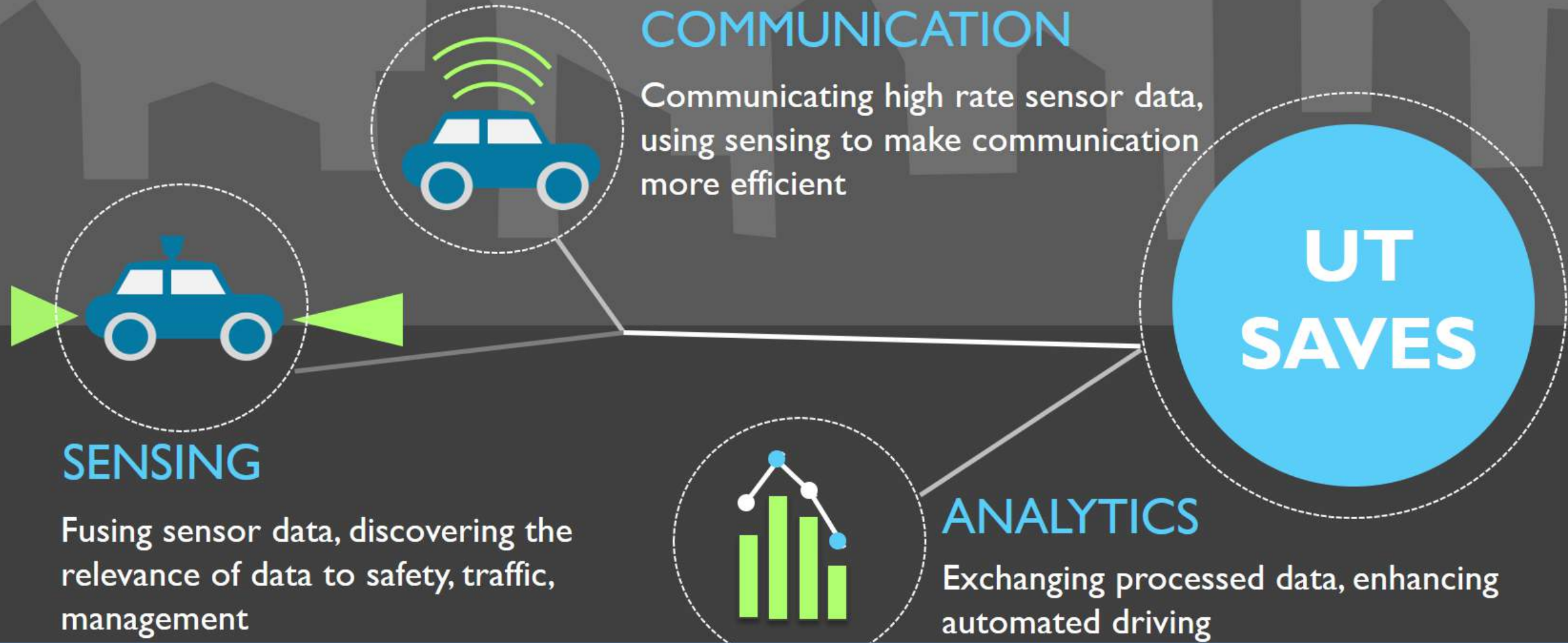


UT SAVES: Situation-Aware Vehicular Engineering Systems





UT SAVES: Situation-Aware Vehicular Engineering Systems



RNL Grand challenge:

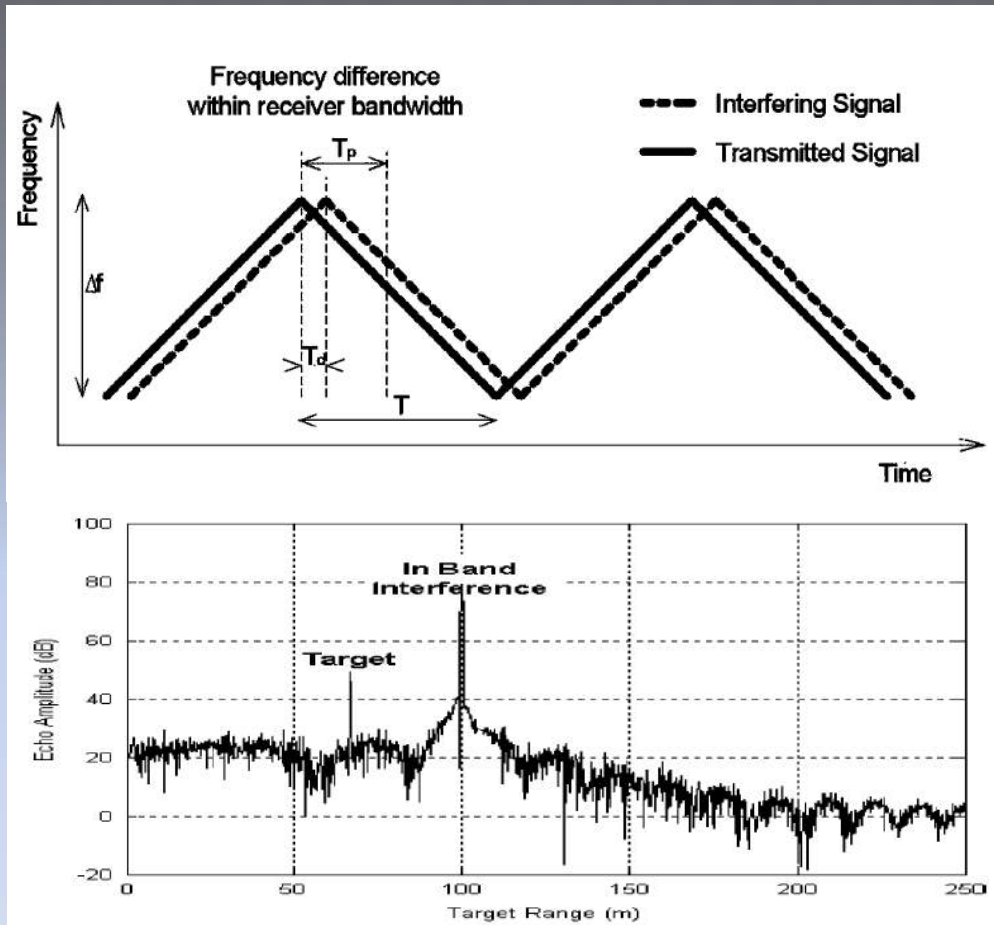
Design automated systems that are both *robust* in the face of unusual natural or accidental events and *secure* against deliberate attack

Robustness and Security at Odds



Improved robustness can reduce security: After the Tesla update to Autopilot 8.x, radar can force automatic braking without cross-validation from camera

Robustness and Security at Odds



Booker, "Mutual interference of mm-wave radar systems." (2007)

The security of automotive radar systems against deliberate attack is weak because the standard FMCW waveform is trivially predictable

Robustness and Security in Unison



Having many subsystems with non-overlapping failure modes improves both robustness and security

**Robust
&
Secure**

Robust & Secure

Robust perception:

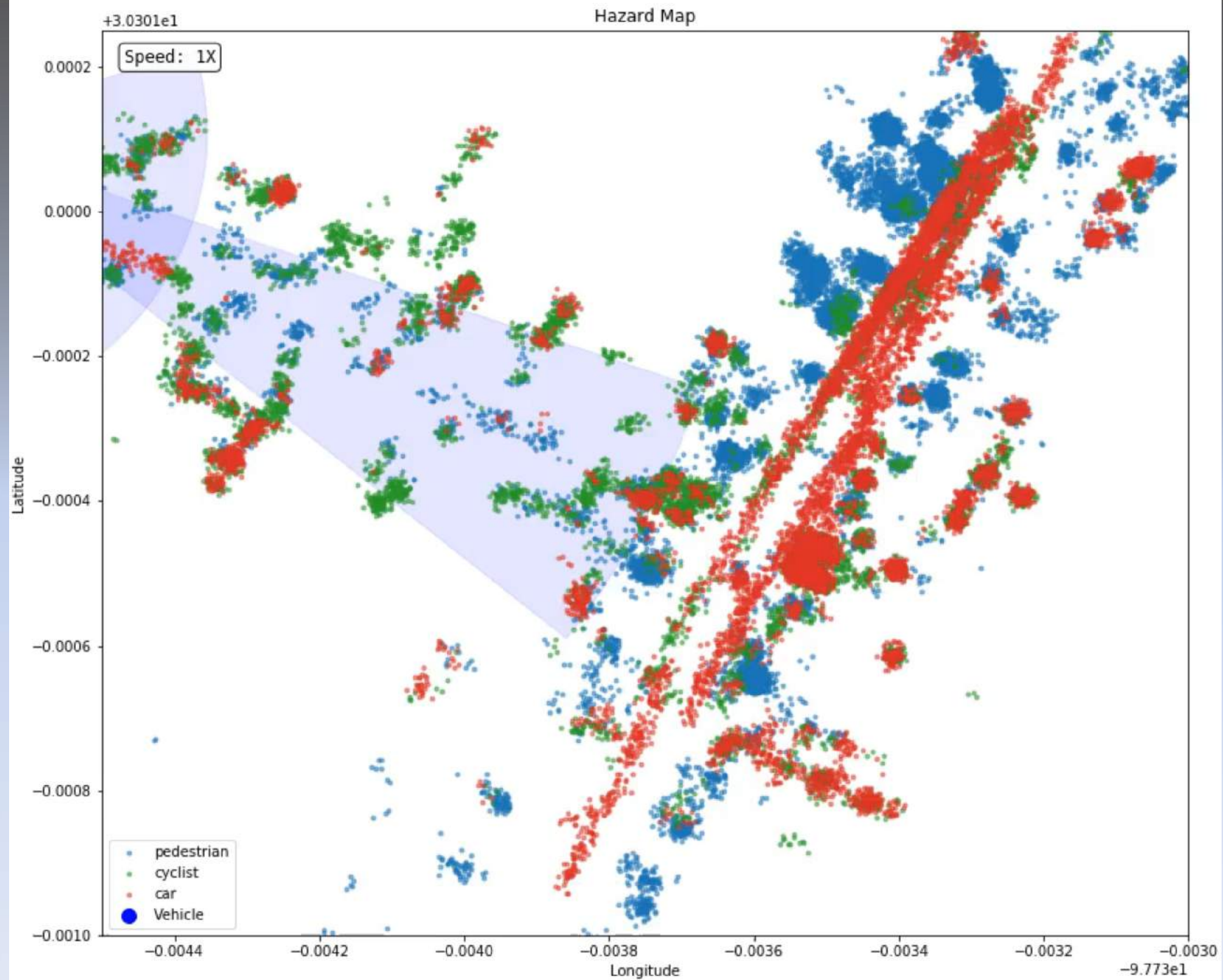
No blind spots

All environments

All weather conditions

**Collaborative
sensing
reduces
risk, increases
perception
robustness**





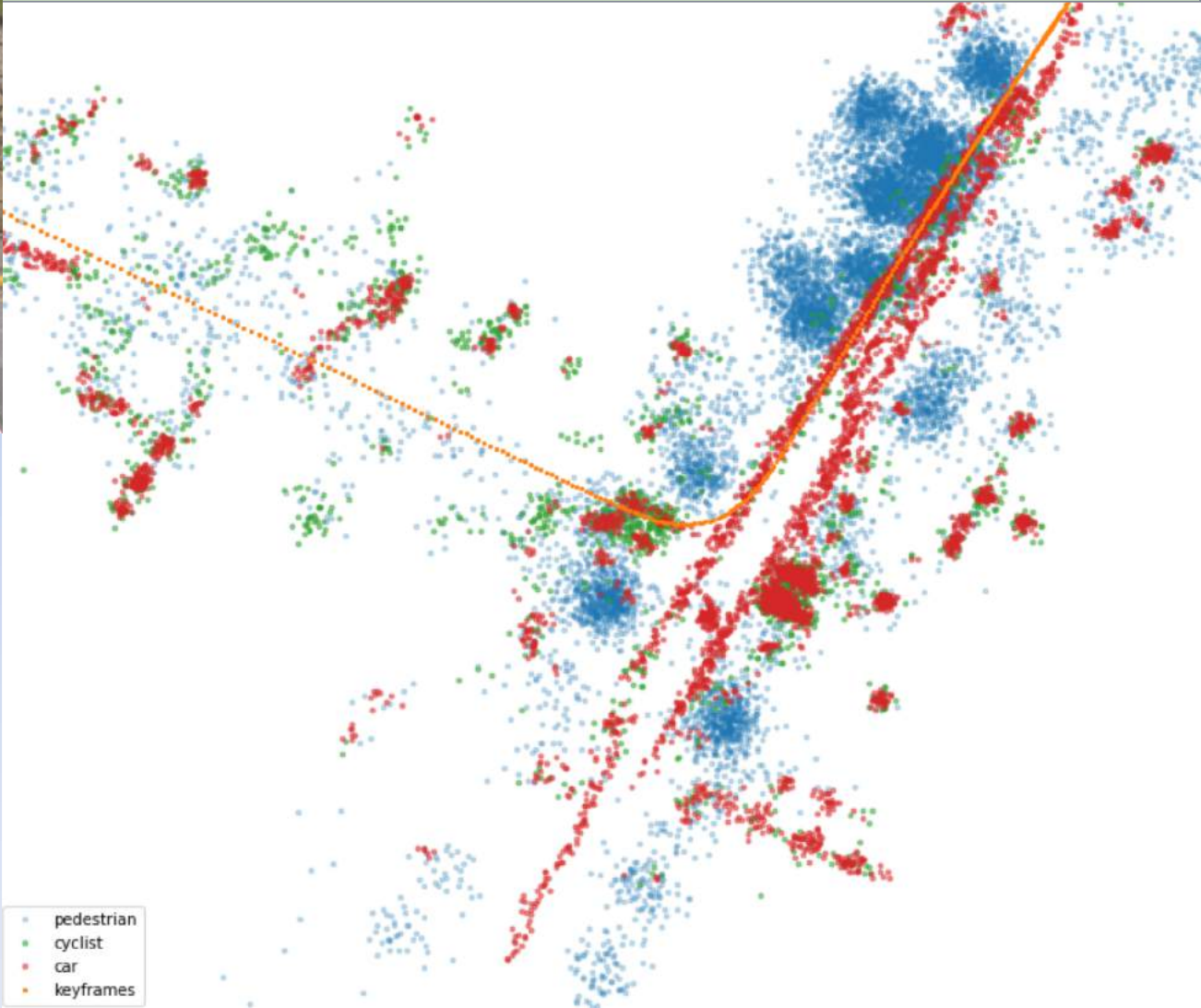
| Features | 5G V2X requirements | LTE C-V2X | DSRC |
|--------------------|--|--------------------------------------|---------------------|
| Data rate | 10 Mb/s @ 1000m 700 Mb/s @ 500m 1 Gb/s @ 50m | Up to 1 Gb/s | 3-27 Mb/s |
| End-to-end Latency | 20 ms , limited auto. 3 ms , full auto. | 20-100 ms | 2-100 ms |
| Reliability | 90~99%, limited auto. 99~99.999%, full auto. | 90% @ 450m* | 90% @ 225m* |
| Range | Up to 1 km | Up to 1km (very large via infra.) | 100 – 1000m |
| Tx Rate | 50 messages/s | Up to 50 messages/s | Up to 20 messages/s |

5G's high data rates and low latency enable multiple vehicles and infrastructure to function as a single sensing organism

| 4 Cameras, 77GHz IF Radar, 77Ghz Delphi Automotive Radar, and GNSS Receiver | | | | |
|---|--|------------------------------|---|--|
| System Sensor | Raw | Processed | Interpreted | |
| Image | PNG 3MP Mono Image - 1 MB | SIFT Image Features - 544kb | Detected Object, Sparse/Dense Reconstructions | |
| Radar | 77GHz FMCW (32 Channels at 80 Mhz) - 8.192 MB | Radar Tracks - 8192 b | Classified Detection and Maps | |
| GNSS | Observables < 1KB | Position/Orientation - 100 B | Ionospheric, Local Multi-path Models | |
| Per Frame Total | 97.53 Mb | 1.1Mb | 1-1000Mb | |
| 10 FPS Total Rate | 975.3 Mb/s | 11 Mb/s | | |



Q: How to quantify reduction in risk for each level of data sharing (raw, processed, interpreted)?



Key insight so far on data sharing:

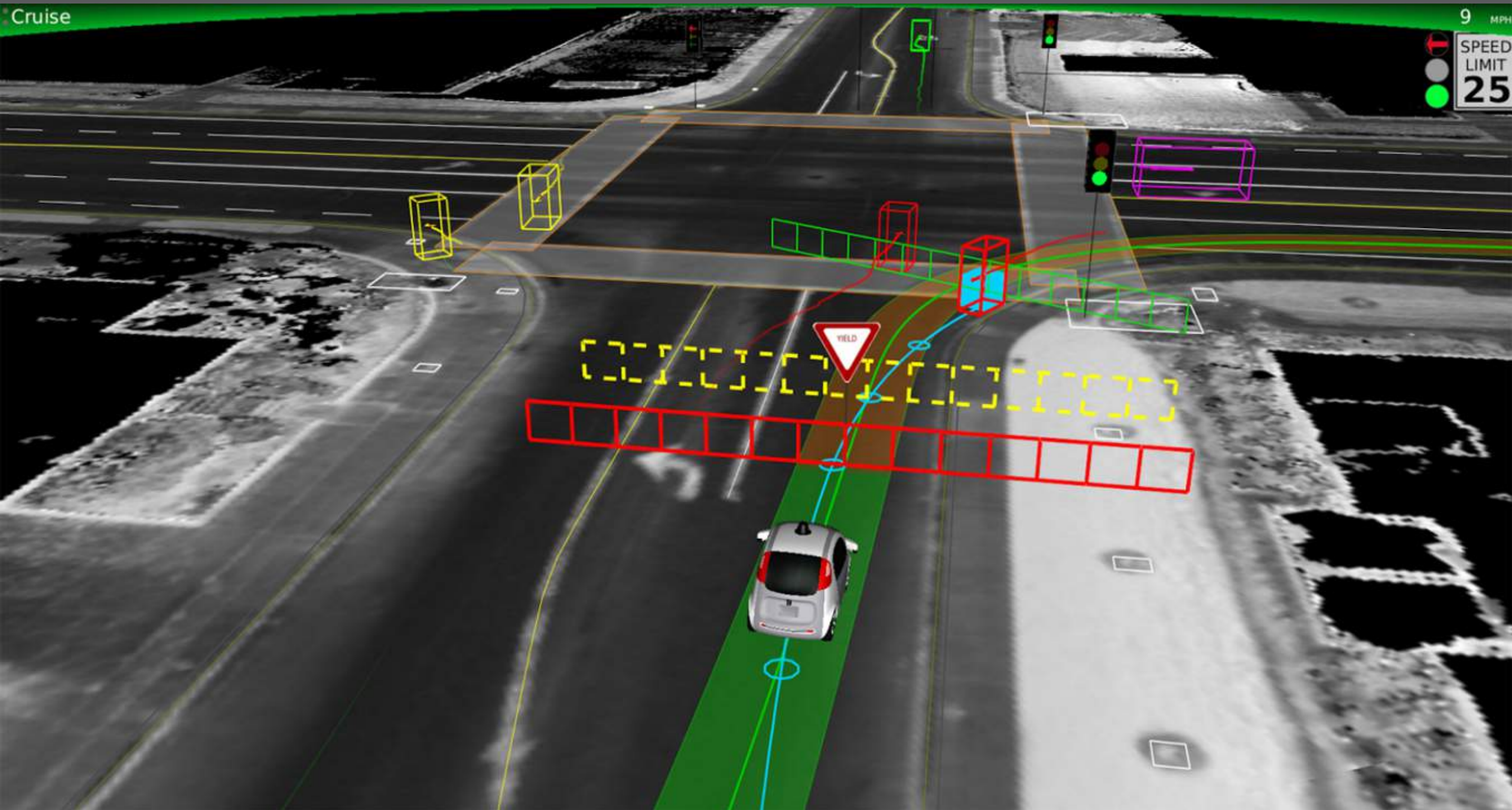
Share either raw data or fully semantically-labeled and located data. Sharing partially-processed data across heterogeneous platforms is less useful

Keys to collaborative all-weather sensing:

- Common reference frame (not just within *your* map)
- Sub-30-cm positioning within common frame
- Decimeter-accurate mapping capability
- Map-merging
- Quick-response and lifetime map maintenance
- Means of *quantifying* the benefit of collaboration

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How accurately does Waymo know the coordinates of this feature in the WGS-84 reference frame?

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In low-visibility conditions, LIDAR and cameras will fail. If pre-mapped radar reflectors are too sparse, lane keeping may be compromised

| SAE level | Name | Narrative Definition | Execution of Steering and Acceleration/Deceleration | Monitoring of Driving Environment | Fallback Performance of Dynamic Driving Task | System Capability (Driving Modes) |
|---|-------------------------------|--|---|-----------------------------------|--|-----------------------------------|
| Human driver monitors the driving environment | | | | | | |
| 0 | No Automation | the full-time performance by the <i>human driver</i> of all aspects of the <i>dynamic driving task</i> , even when enhanced by warning or intervention systems | Human driver | Human driver | Human driver | n/a |
| 1 | Driver Assistance | the <i>driving mode</i> -specific execution by a driver assistance system of either steering or acceleration/deceleration using information about the driving environment and with the expectation that the <i>human driver</i> perform all remaining aspects of the <i>dynamic driving task</i> | Human driver and system | Human driver | Human driver | Some driving modes |
| 2 | Partial Automation | the <i>driving mode</i> -specific execution by one or more driver assistance systems of both steering and acceleration/deceleration using information about the driving environment and with the expectation that the <i>human driver</i> perform all remaining aspects of the <i>dynamic driving task</i> | System | Human driver | Human driver | Some driving modes |
| Automated driving system ("system") monitors the driving environment | | | | | | |
| 3 | Conditional Automation | the <i>driving mode</i> -specific performance by an <i>automated driving system</i> of all aspects of the dynamic driving task with the expectation that the <i>human driver</i> will respond appropriately to a <i>request to intervene</i> | System | System | Human driver | Some driving modes |
| 4 | High Automation | the <i>driving mode</i> -specific performance by an automated driving system of all aspects of the <i>dynamic driving task</i> , even if a <i>human driver</i> does not respond appropriately to a <i>request to intervene</i> | System | System | System | Some driving modes |
| 5 | Full Automation | the full-time performance by an <i>automated driving system</i> of all aspects of the <i>dynamic driving task</i> under all roadway and environmental conditions that can be managed by a <i>human driver</i> | System | System | System | All driving modes |

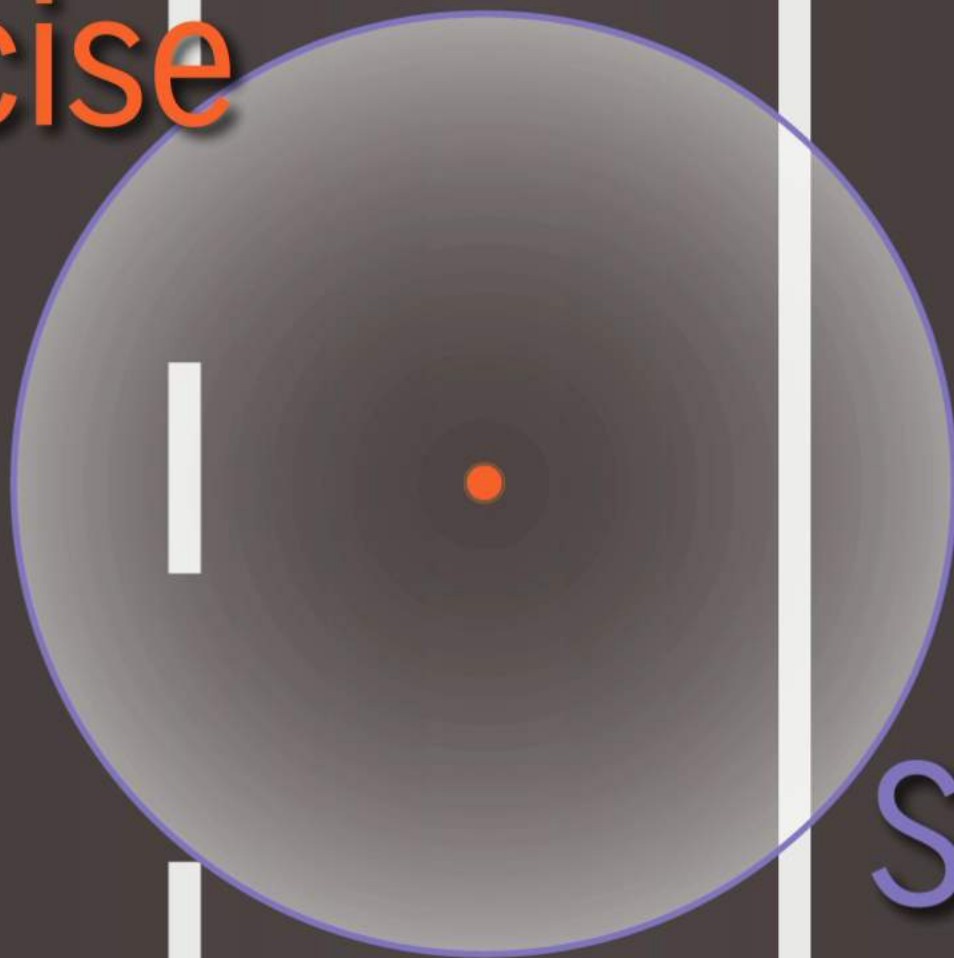
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Call to robustness:

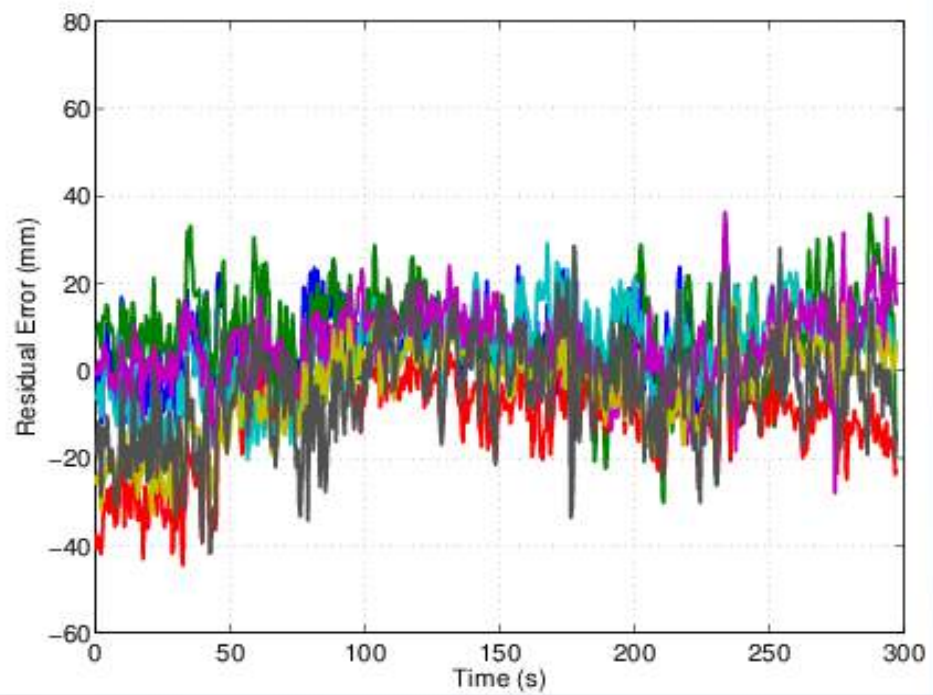
**We should be designing for Level 6,
Continued operation under conditions in
which a human couldn't manage**



Precise

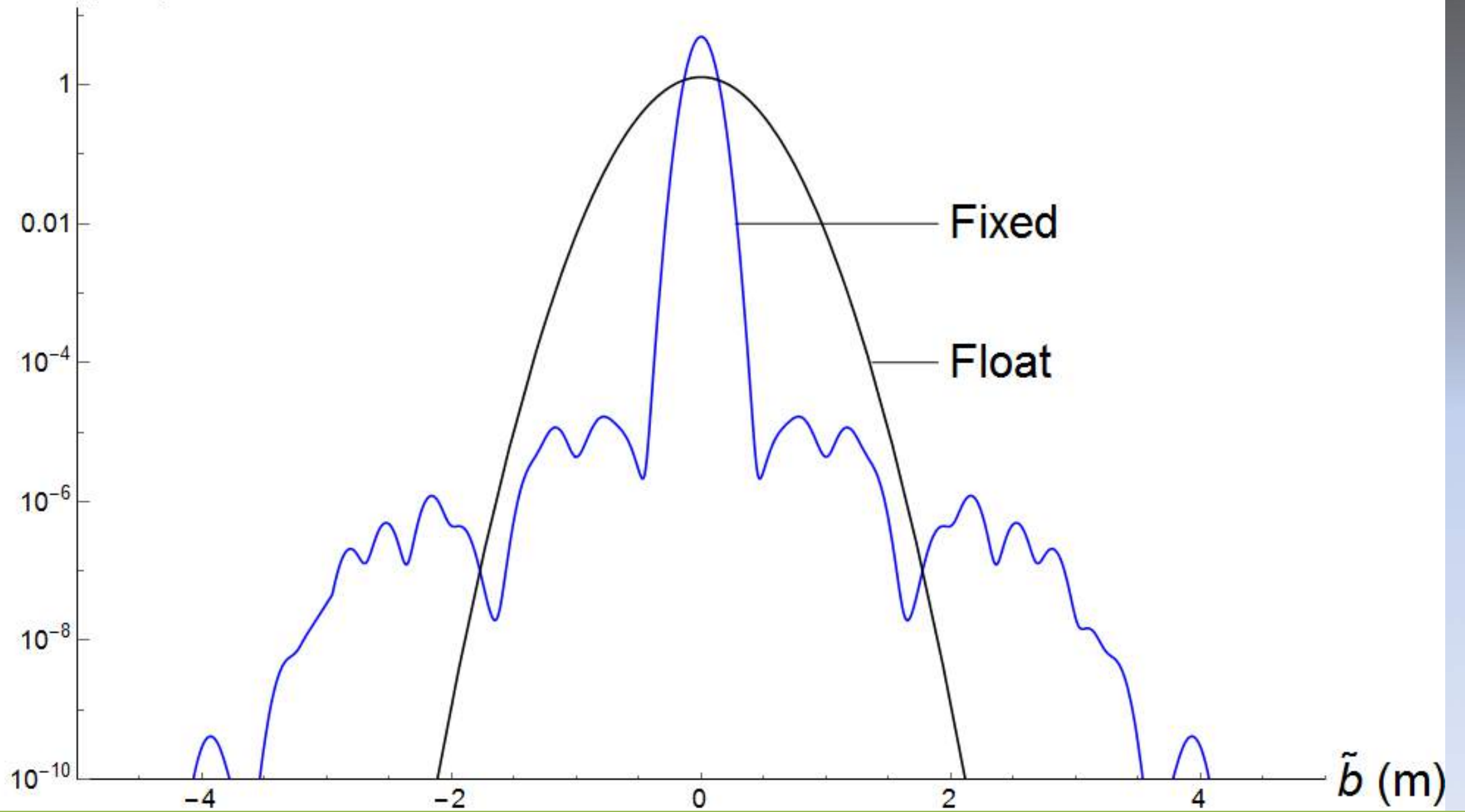


Standard



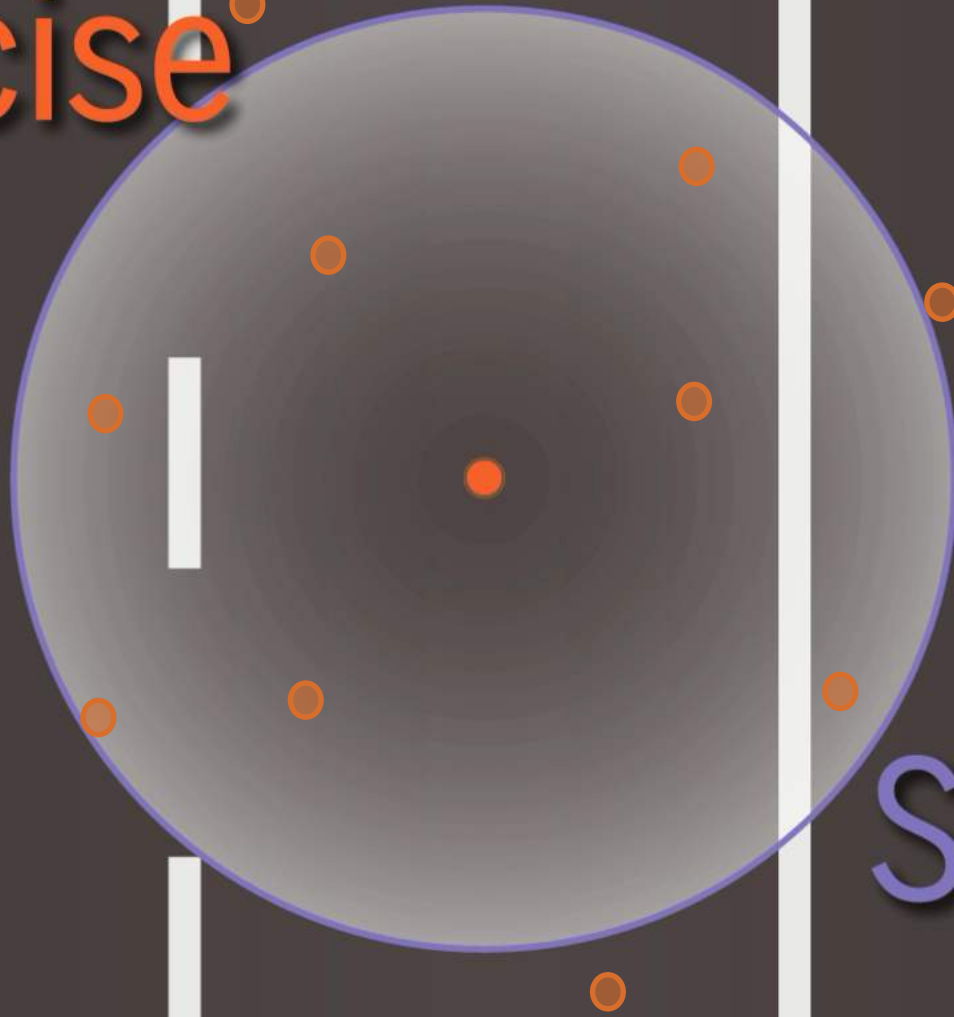
Dec. 2014: First successful RTK solution with a smartphone antenna.

PDF (log)



Float solution does not enforce integer constraint. Fixed solution does, but may be wrong.

Precise



Standard

Safety-of-life applications require careful management of incorrect fixing probability

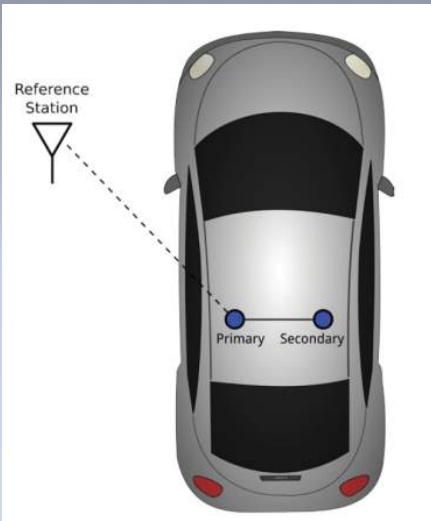
Light urban setting



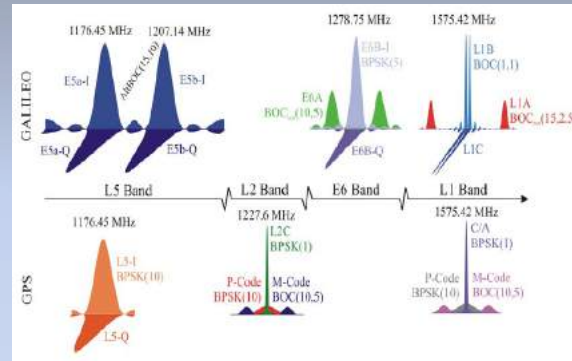
Urban setting



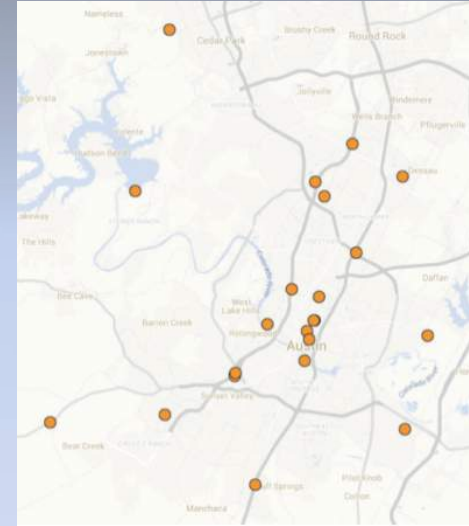
Keys to Robust CDGNSS Positioning



Multiple Rover
Antennas



Multiple
Frequencies

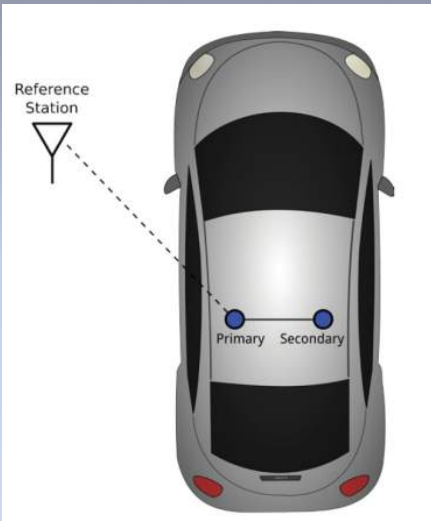


Dense Reference
Network

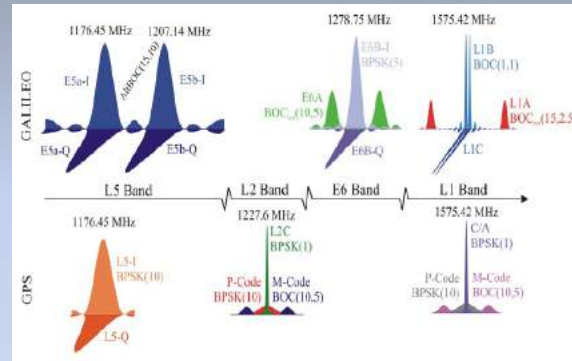


Fusion with
Vision/Radar

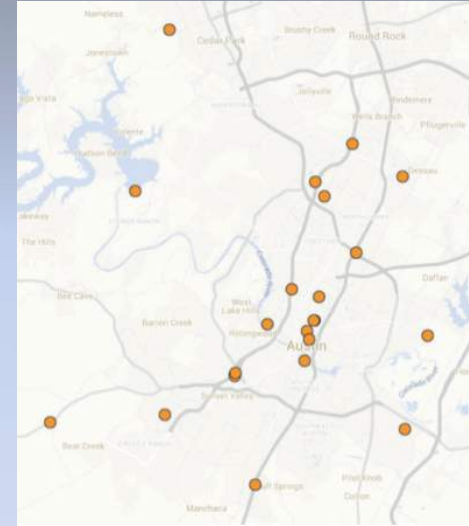
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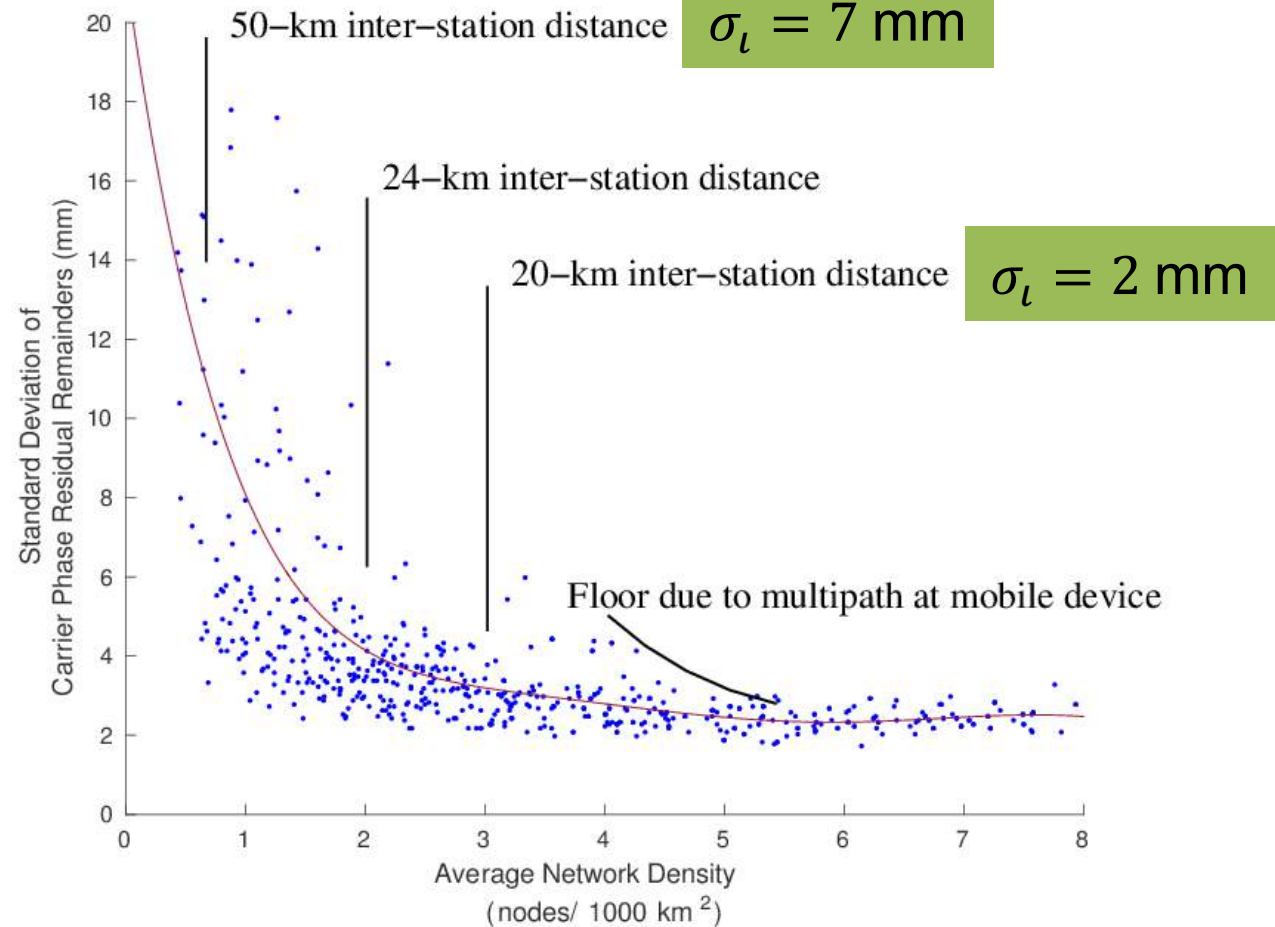
Dense Reference
Network



Fusion with
Vision/Radar

Don't waste valuable signals estimating atmospheric errors: you'll need them to combat your chief adversary: multipath

Eliminate iono with a dense reference network



Corrections uncertainty is a highly nonlinear function of density. Culprit: medium-scale ionospheric irregularities. **Floor due to multipath at mobile device reached with < 20km distance between reference stations.**



Longhorn Dense Reference Network stations under test prior to deployment



UT SAVES Sensorium v2.0:

Stereo cameras, dual-antenna triple-frequency software-defined GNSS, industrial-grade IMU (8 deg/hr gyros), automotive radar, LTE connectivity



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Stereo cameras, dual-antenna triple-frequency software-defined GNSS, industrial-grade IMU (8 deg/hr gyros), automotive radar, LTE connectivity



Legend

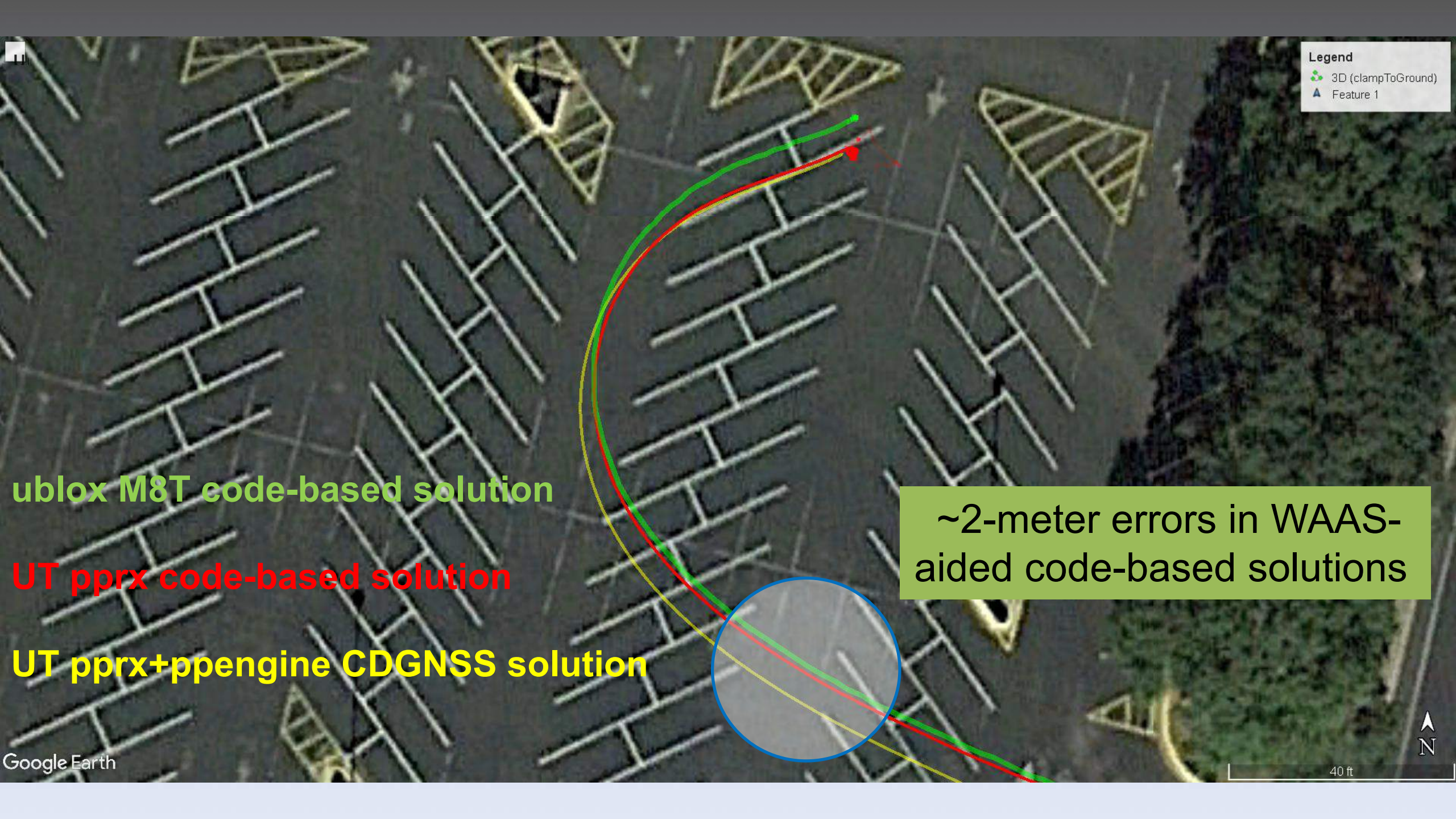
▲ sps

Google Earth

© 2008 Sanborn

200 ft

N



Legend

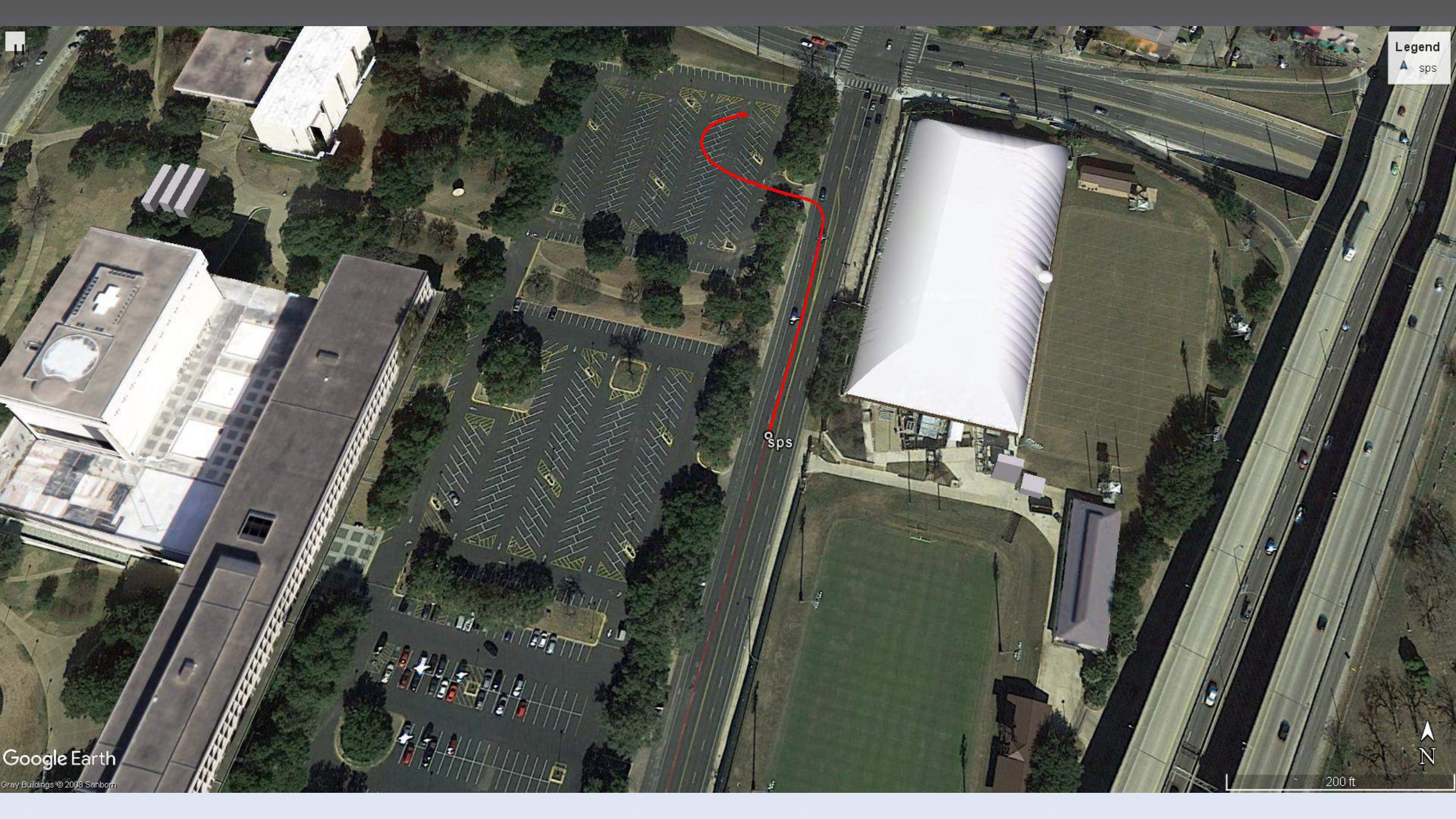
- 3D (clampToGround)
- Feature 1

ublox M8T code-based solution

UT pprx code-based solution

UT pprx+ppengine CDGNSS solution

~2-meter errors in WAAS-aided code-based solutions



Legend

▲ sps

Google Earth

Gray Buildings © 2008 Sanborn

200 ft





Legend

- 3D (clampToGround)
- Feature 1

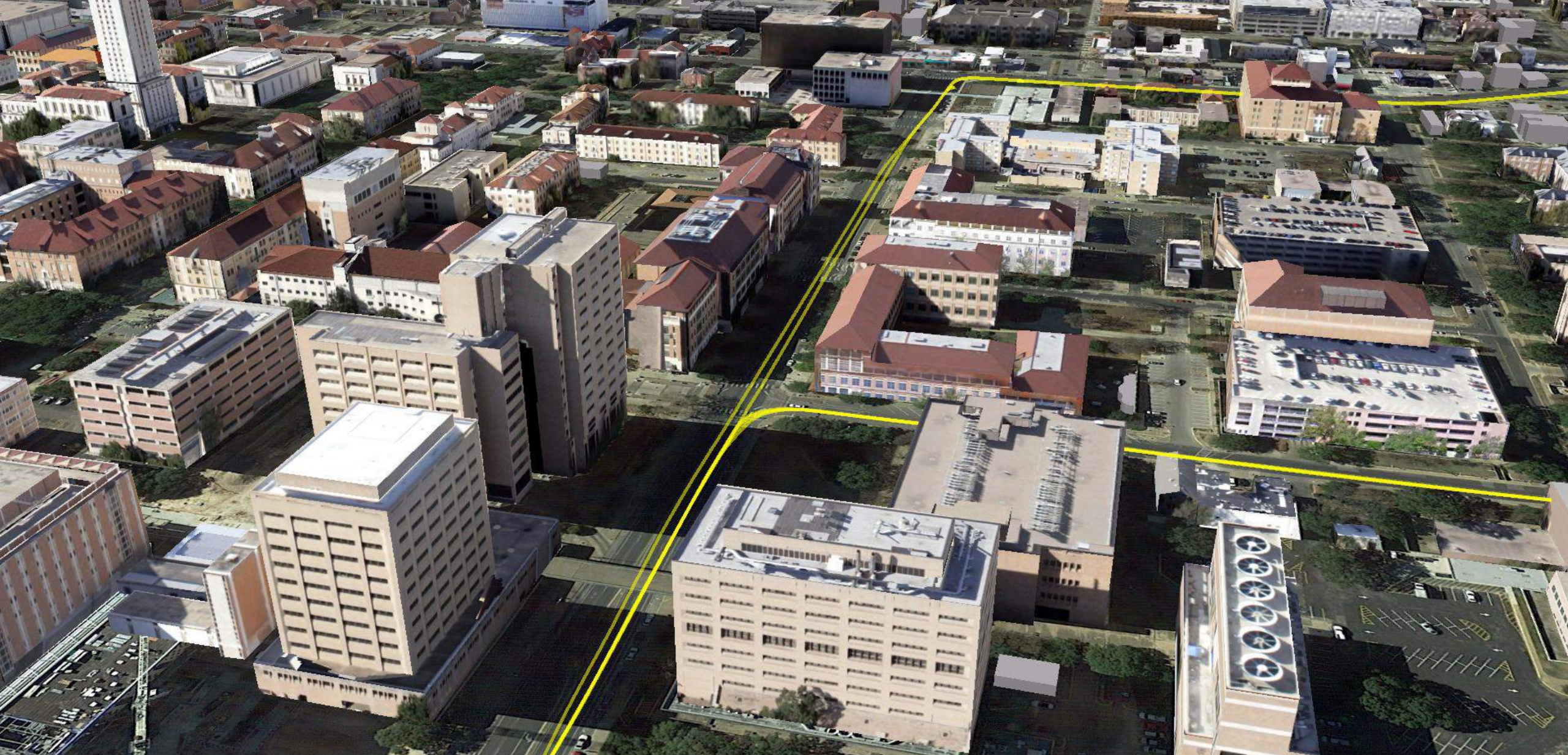
ublox M8T code-based solution

UT pprx code-based solution

UT pprx+ppengine CDGNSS solution



January 2018 test of single-antenna stand-alone precise GNSS in moderate urban environment without aiding, inertial or otherwise

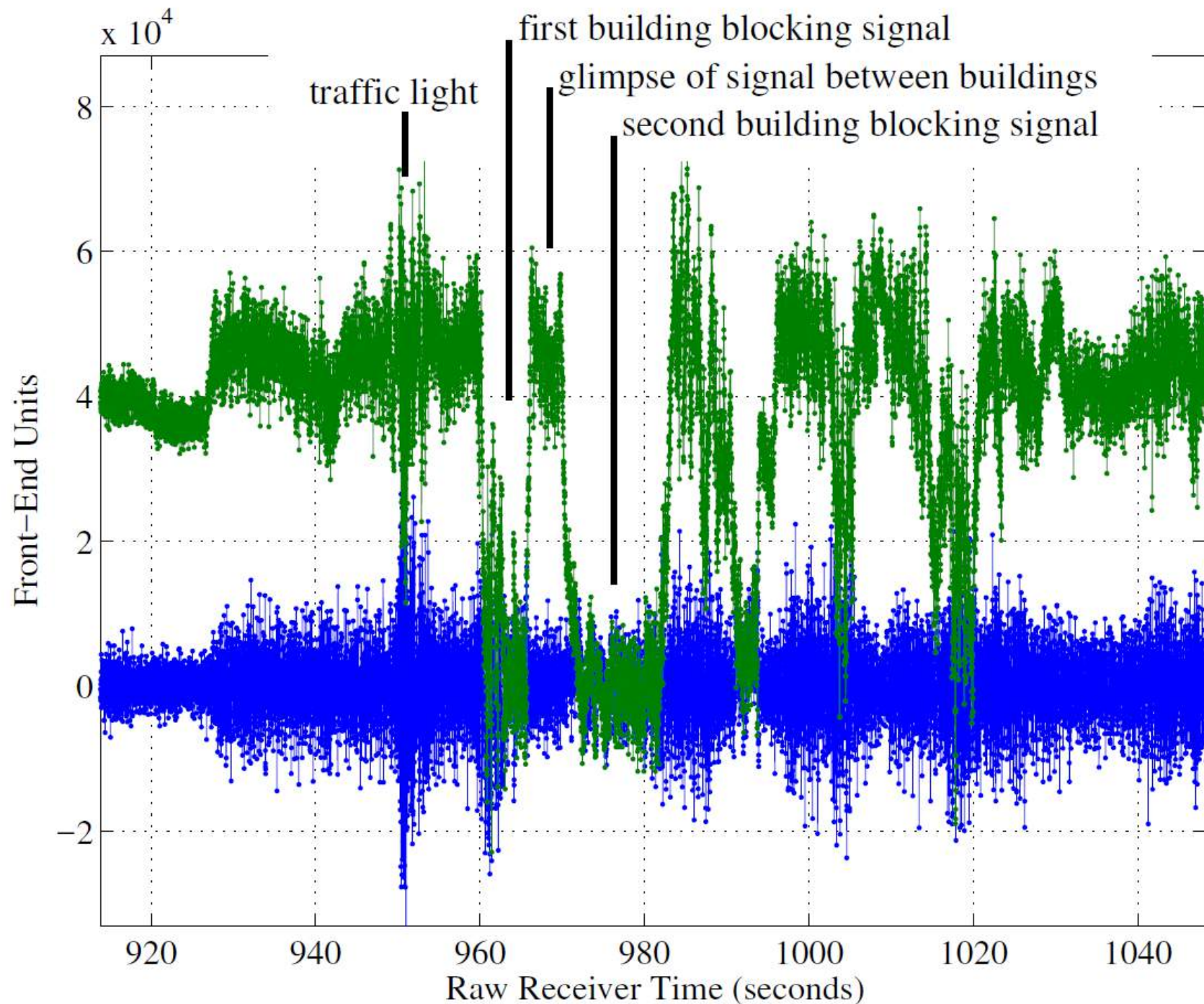


Dean Keaton corridor:
Two bridges, overhanging trees, tall buildings on both sides of street



Dean Keaton corridor:
Two bridges, overhanging trees, tall buildings on both sides of street

Carrier tracking must
be fast and
opportunistic but
avoid frequency
unlock



Promising results:

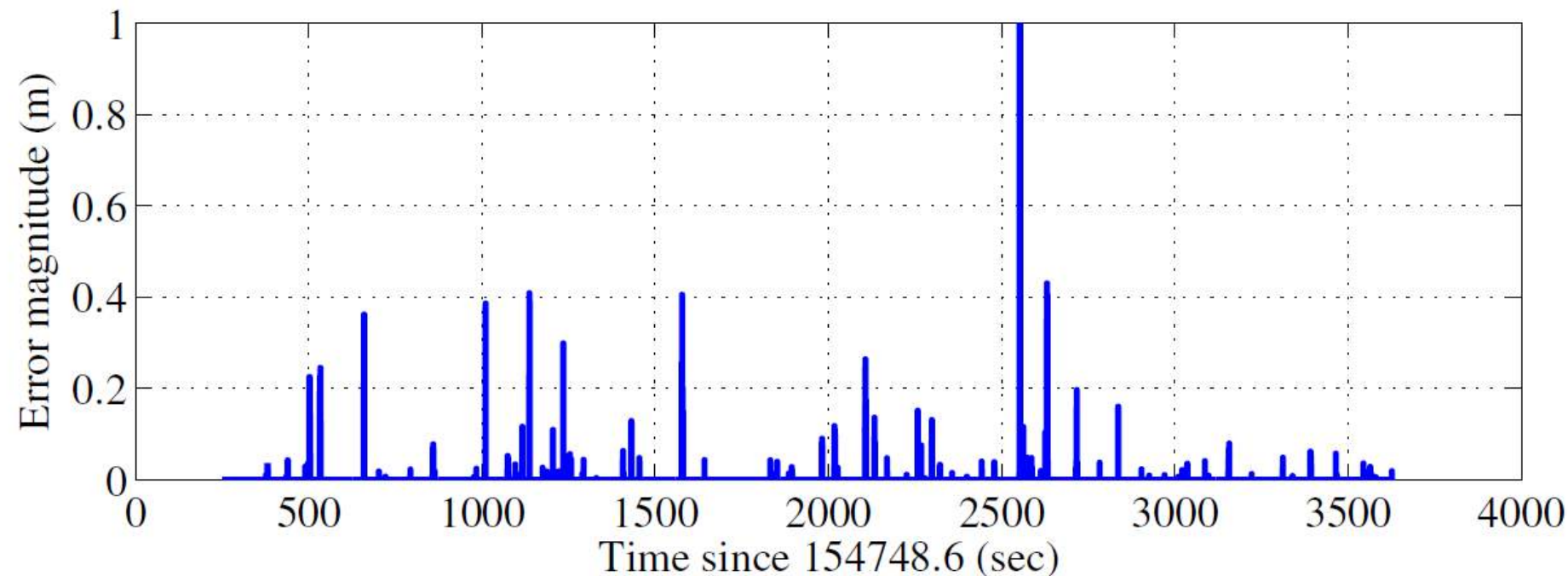
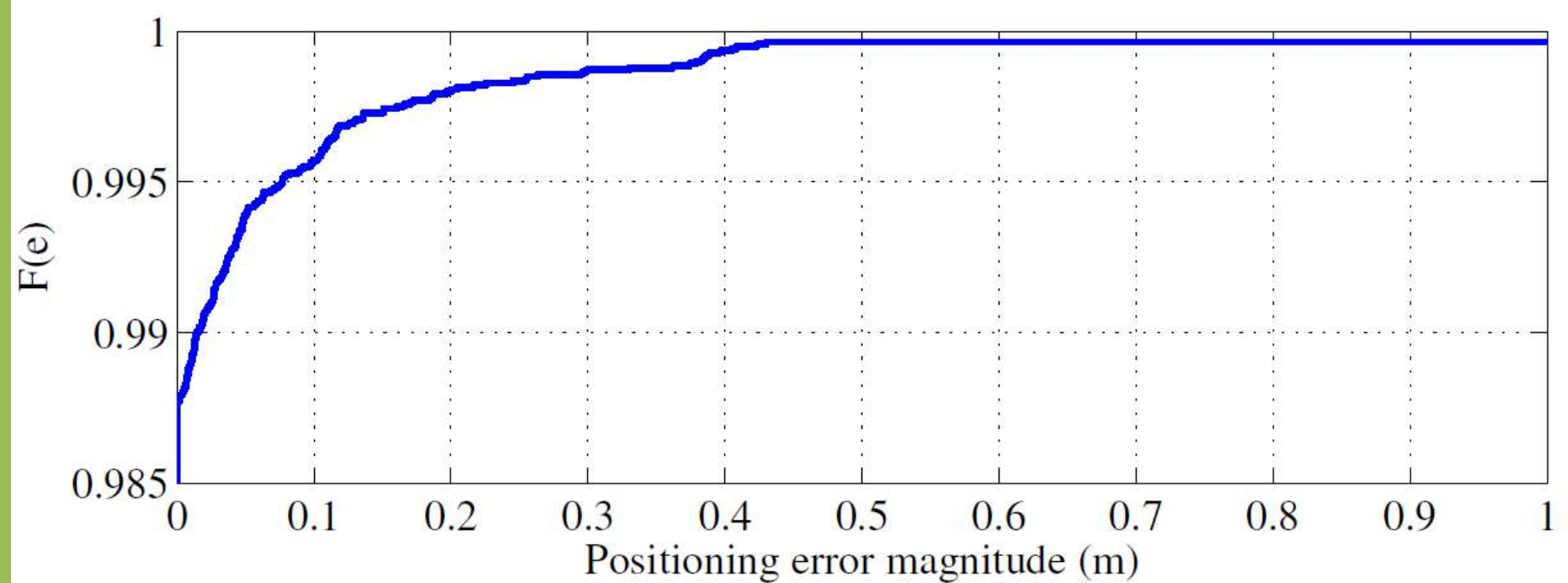
90% availability of
validated solutions in
1 hour of driving

Validated solutions
accurate to 10 cm for
>99.5% of testable
epochs

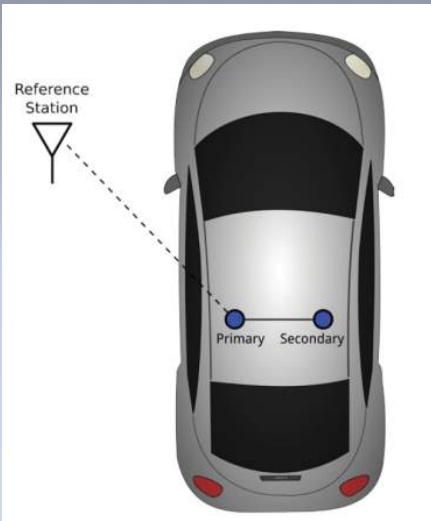
Still not good enough:

Need 100% avail.

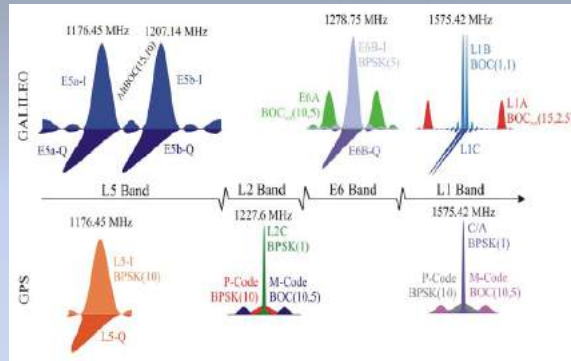
Even 0.2% errors
greater than 30 cm is
unacceptable



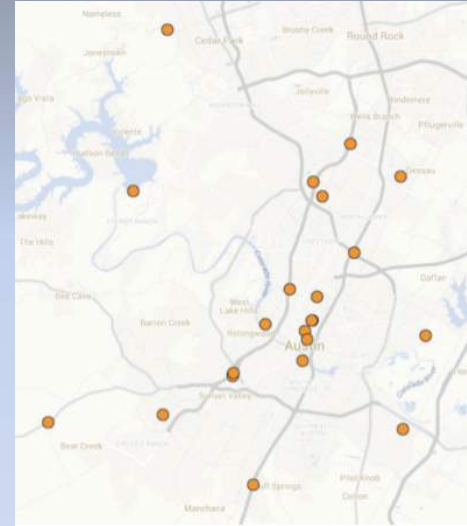
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Dense Reference
Network



Fusion with
Vision/Radar



Goal: Decimeter-accurate visual and radar features



GNSS-INS provides initial pose and global reference



Merges globally-referenced maps from multiple sessions



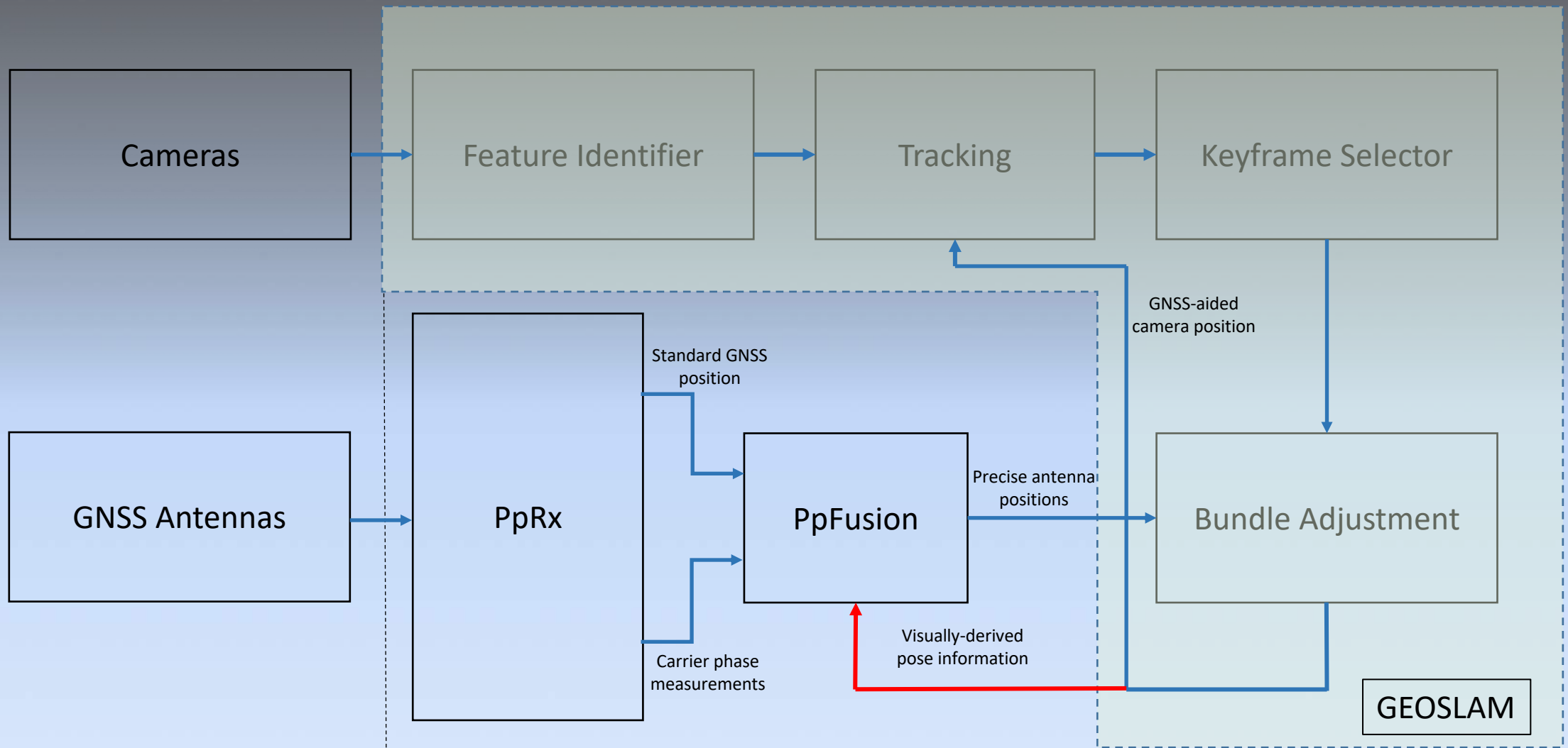
Cooperative fleet refinement of map



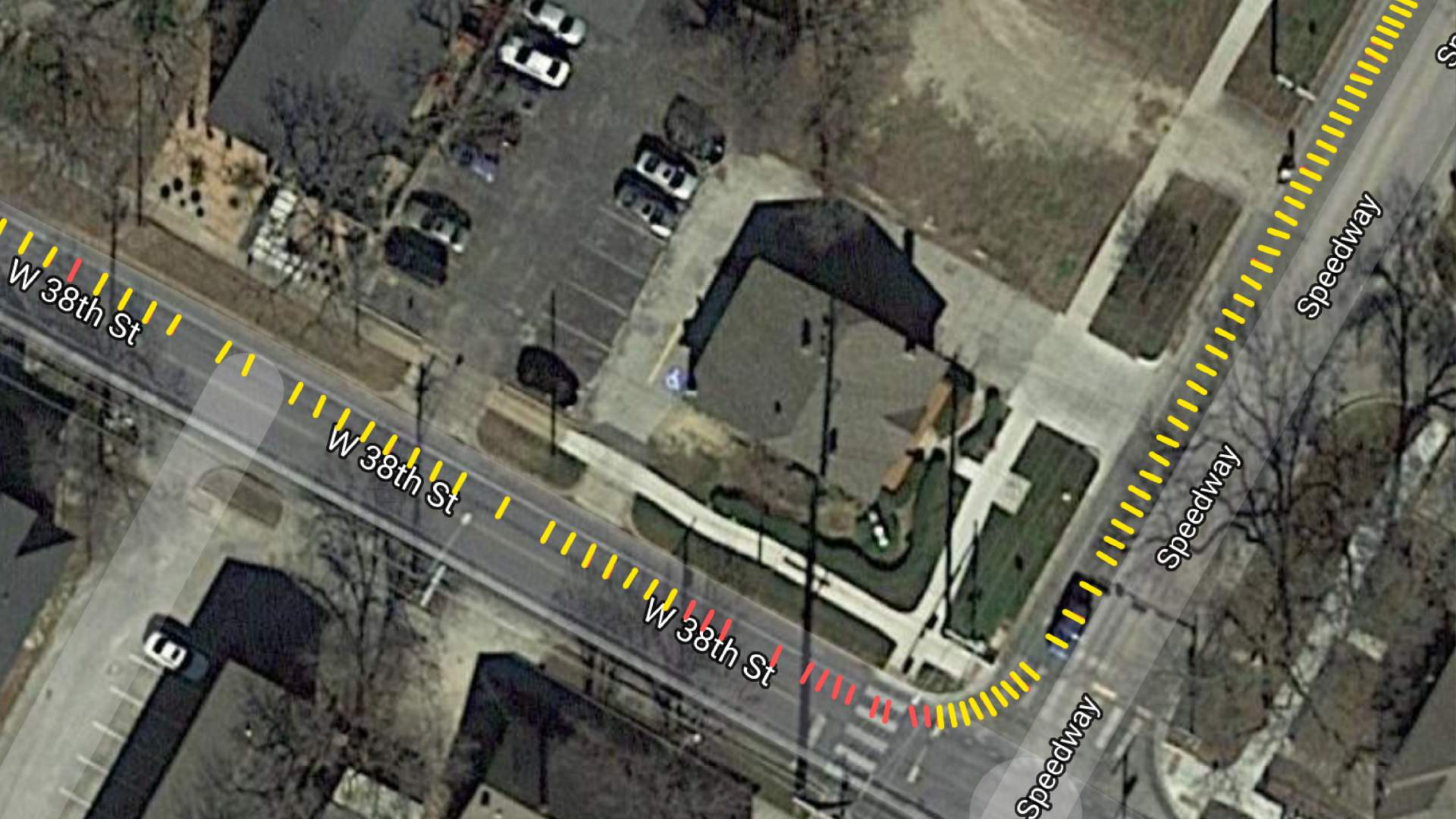
GNSS Antennas
Stereo Cameras
Radar
Inertial Sensor

GEOSLAM: Global Electro-Optical SLAM

Cooperative mapping at decimeter accuracy for all-weather localization



Tightly-coupled CDGNSS-Vision system





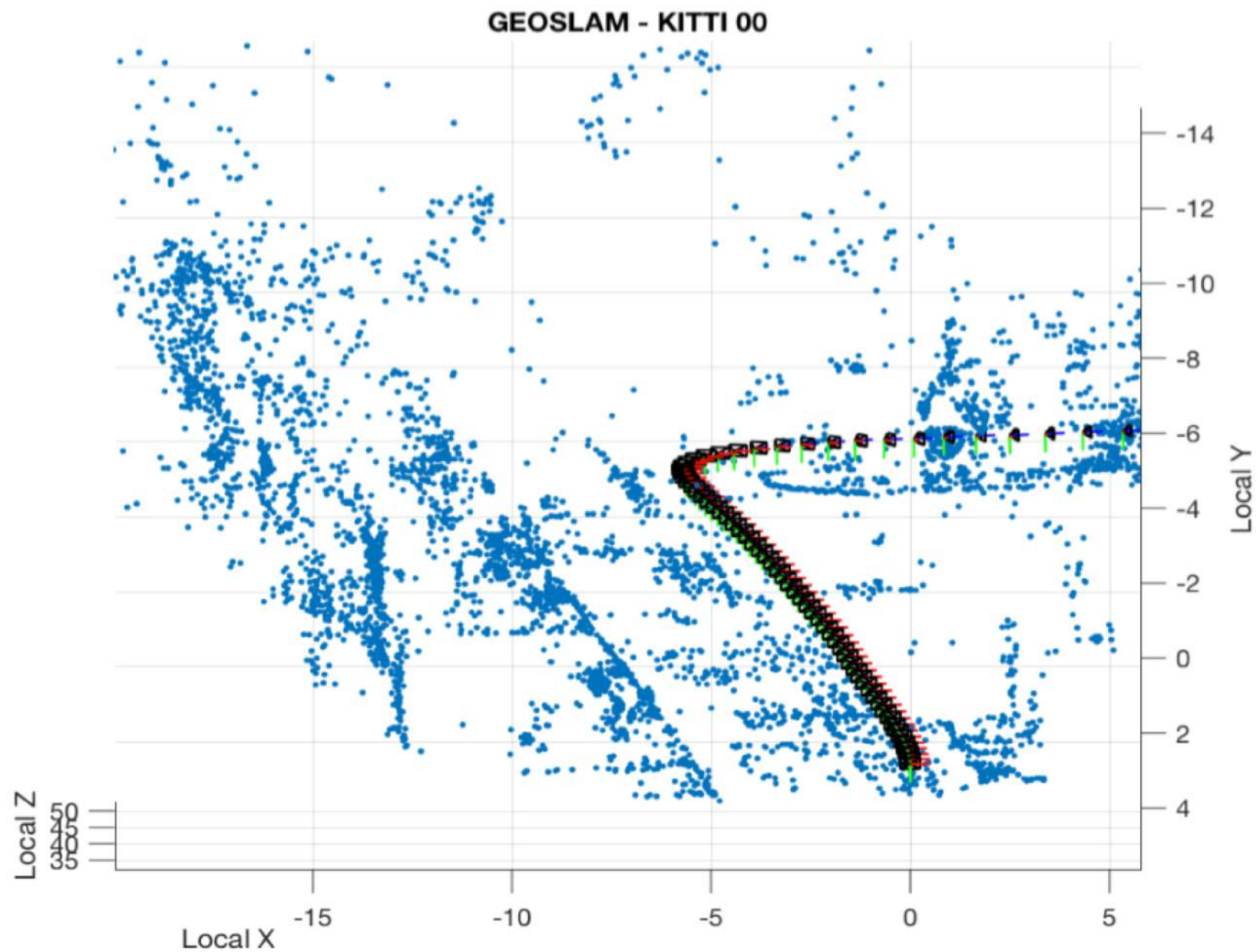
W 38th St

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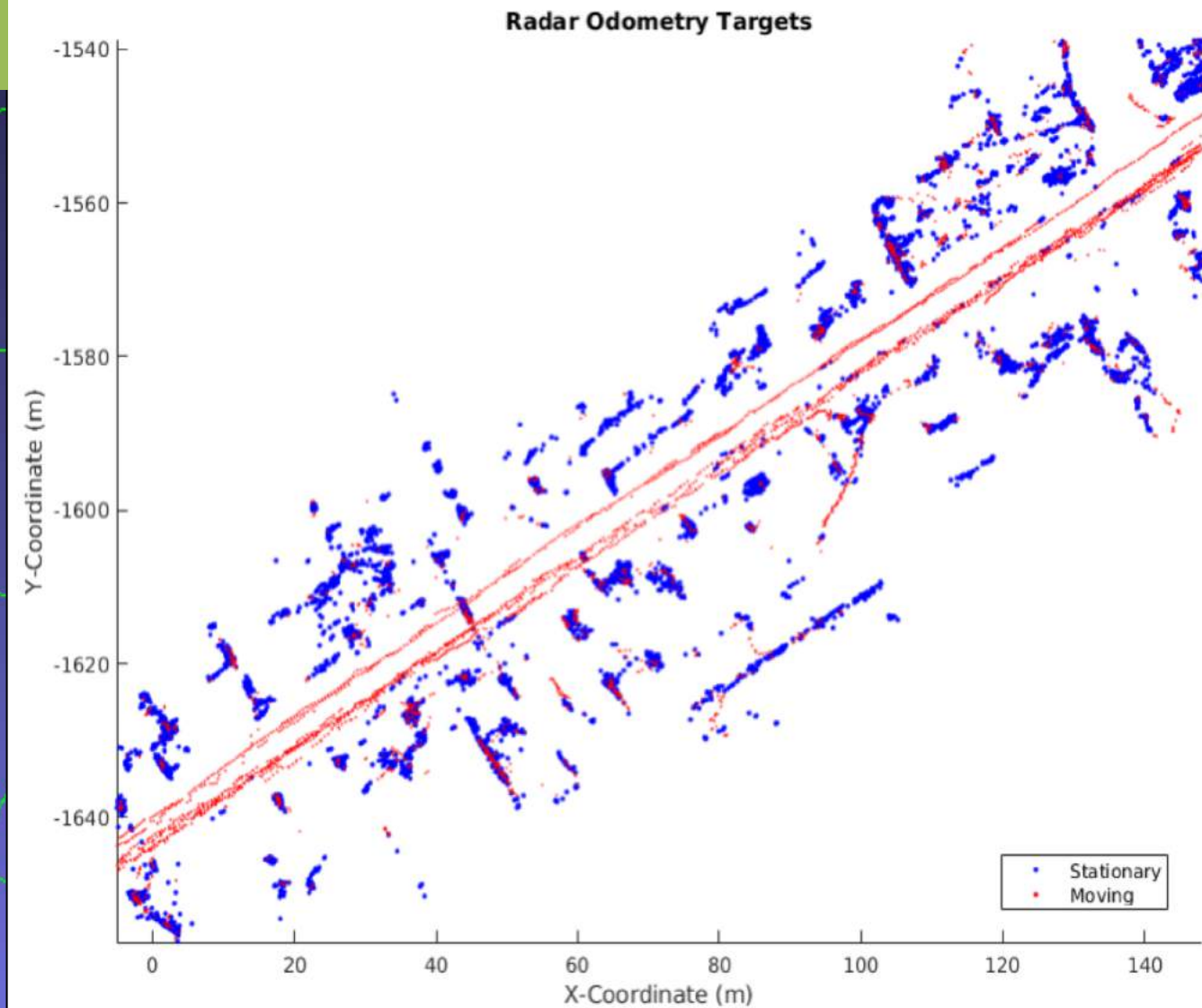
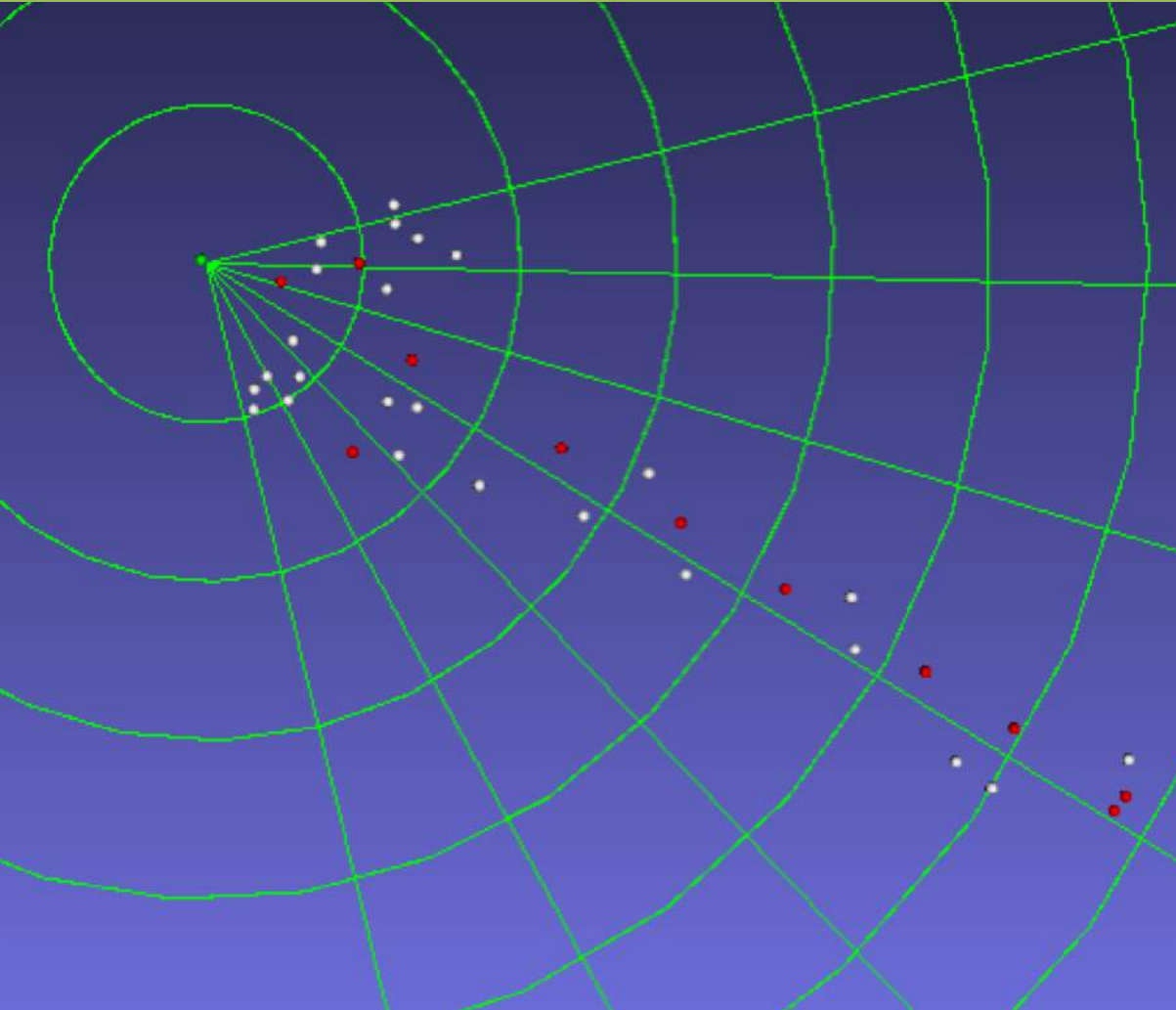
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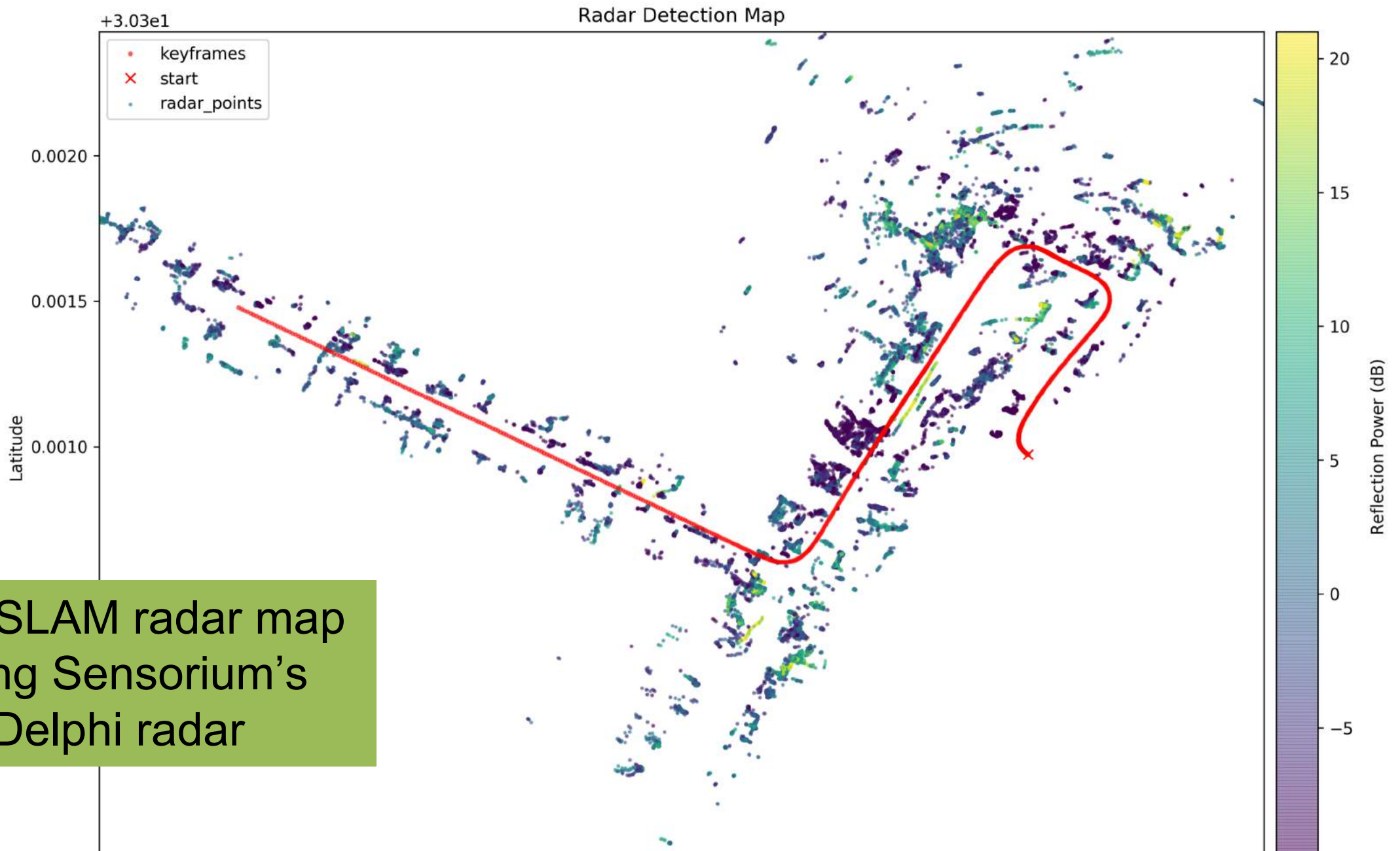
3D Map from GEOSLAM test on KITTI data set

Map of 3D camera image features is a means to an end: it can't be relied on when visibility poor, but when visibility good, it can be used (jointly with GNSS) to obtain highly accurate vehicle pose with which a radar map can be built

Sensorium radar data: Delphi unit



A radar map is useful in all weather conditions.
Our conjecture: In urban areas rich with radar reflectors, a radar map together with GNSS will be sufficient for 30 cm @ 95% accuracy



GEOSLAM radar map
using Sensorium's
Delphi radar

Goal: 30-cm all-environment positioning

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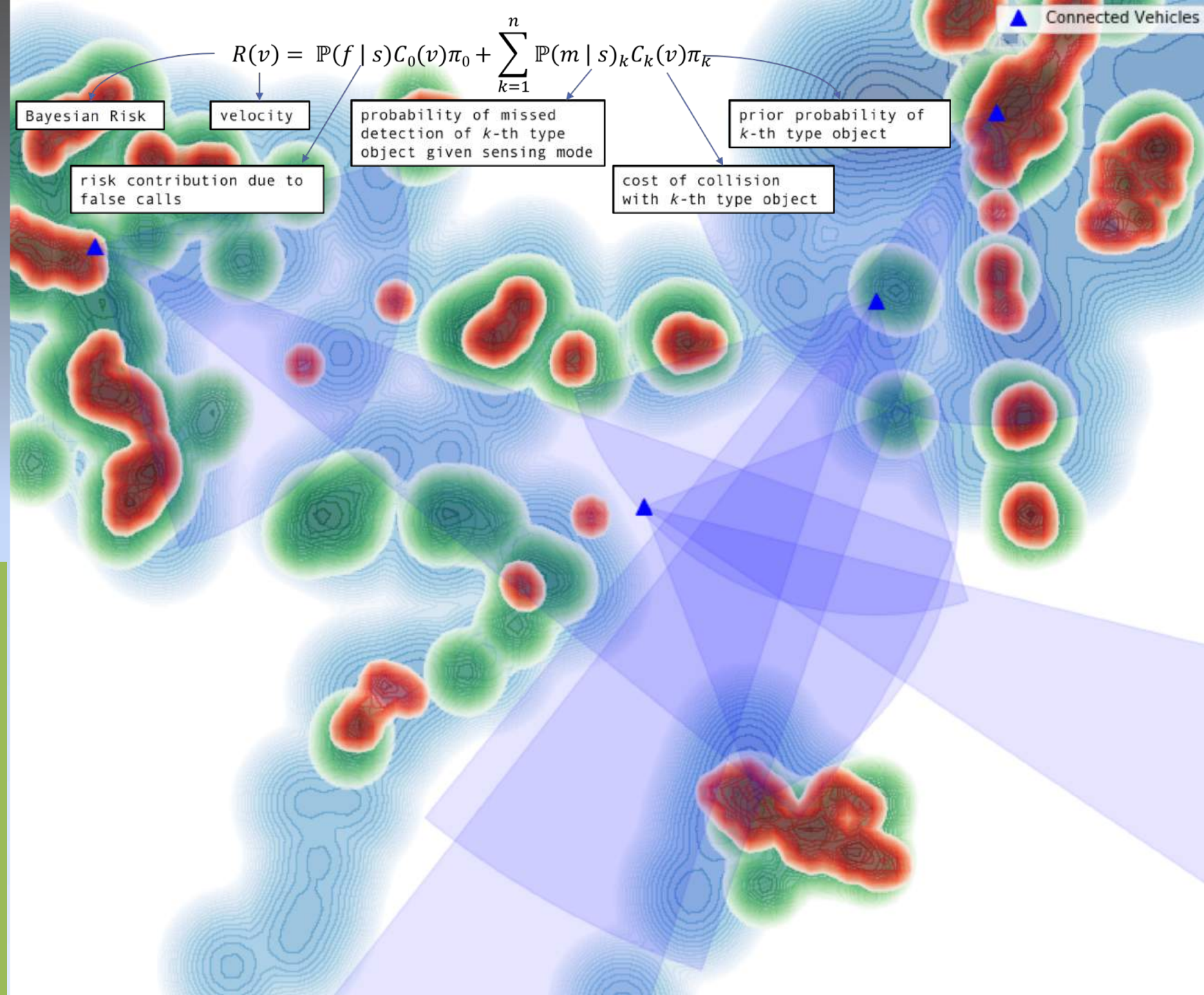
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We take a Bayesian approach to risk evaluation

Occupancy distributions for pedestrians, cyclists, and vehicles indicate risk of hazardous interactions. As multiple vehicles share sensor data, the risk is reduced







Resulting 3D model has mm-level resolution and sub-cm absolute accuracy
Research goal: Multi-UAV fully-automated collaborative high-resolution mapping

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